

Advances in Parallelizing Algebraic Multigrid

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The BoomerAMG Team

- Algorithm development, design, codewriting, maintenance
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- Consulting
 - John Ruge

Boomer AMG

Outline

- Parallelization of AMG
- Coarsening techniques
- Relaxation techniques
- Numerical results
- Conclusions



AMG has two phases:

Setup Phase

- Select Coarse "grids," Ω^{m+1} , $m=1,2,\ldots$
- Define interpolation, I_{m+1}^m , m = 1, 2, ...
- Define restriction and coarse-grid operators

$$I_m^{m+1} = (I_{m+1}^m)^T$$
 $A^{m+1} = I_m^{m+1} A^m I_{m+1}^m$

Solve Phase

Standard multigrid operations, e.g., V-cycle, W-cycle, FMG, etc

We must parallelize these steps:



- In The Setup Phase
 - Coarse Grid Selection
 - Construction of Prolongation operator, P
 - Construction of coarse-grid operators by Galerkin method, RAP, R=P'
- In The Solve Phase
 - Residual Calculation
 - Relaxation
 - Prolongation
 - Restriction



Parallelizing the Solve Phase

- In The Solve Phase
 - Residual Calculation
 - entails Axpy Matvec: y<-aAx+by.</p>
 - Relaxation
 - Jacobi is essentially a Matvec
 - Gauß-Seidel is sequential, but hybrid (or chaotic) schemes may be employed
 - Prolongation
 - requires a Matvec (on a rectangular matrix)
 - Restriction
 - requires a MatvecT

Basic concept: Smooth error means "small" residuals



Error that is slow to converge obeys:

$$e^{k+1} = (I - Q^{-1}A) e^k$$
; hence $(I - Q^{-1}A) e \approx e$
 $\Rightarrow Q^{-1}A e \approx 0 \Rightarrow r \approx 0$

Define: i depends on j (and j influences i) if

$$-a_{ij} \ge \theta \max_{k \ne i} \{-a_{ik}\}, \quad 0 < \theta \le 1$$

The set of dependencies of i is given by

$$S_{i} = \left\{ j : -a_{ij} > \theta \max_{j \neq i} -a_{ij} \right\}$$

Smooth error varies slowly in the direction of dependence



Coarsening techniques

- classical Ruge- Stüben(RS) algorithm
- Cleary-Luby-Jones-Plassman (CLJP) algorithm
- parallel Ruge-Stüben coarsening techniques
- Falgout-CLJP coarsening



Choosing the Coarse Grid

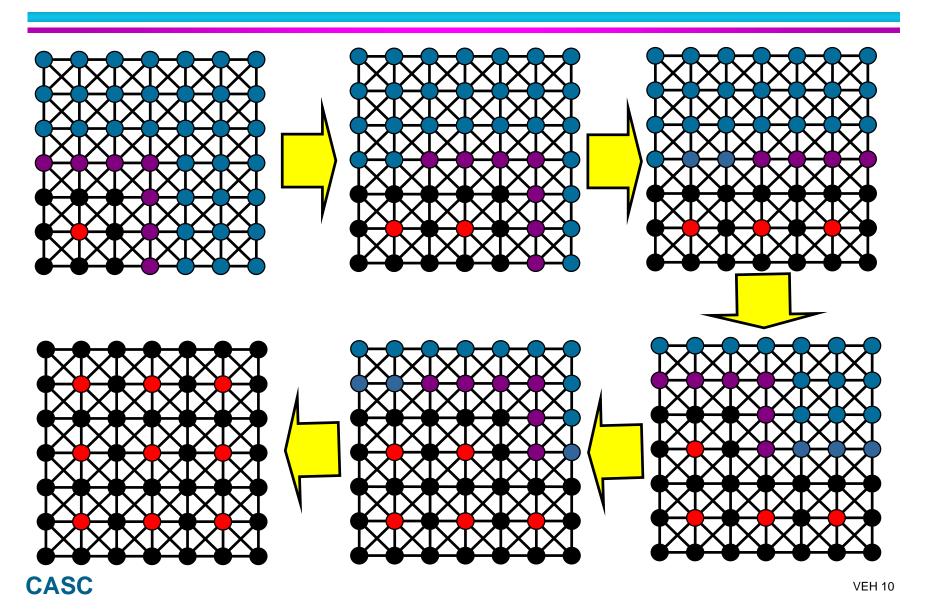
Two Criteria

- (C1) For each $i \in F$, each point $j \in S_i$ should either be in C or should be strongly connected to at least one point in C_i
- (C2) C should be a maximal subset with the property that no two C-points are strongly connected to each other.
- Satisfying both (C1) and (C2) is sometimes impossible. We use (C2) as a guide while enforcing (C1).

CASC

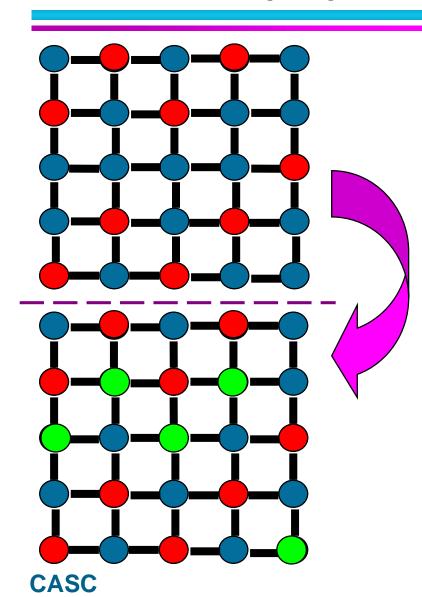
The AMG coarse-grid selection algorithm is inherently sequential





A second pass is needed to enforce (C1)





 First-pass coarsening of 5 point Laplacian, periodic boundary conditions

 Numerous F-F dependencies among points not sharing common C-point

 A second "coloring" pass is made, changing F-points to Cpoints, as needed, to ensure (C1).

Parallel Ruge-Stüben Coarsening



- One approach to coarsening in parallel: perform the standard Ruge- Stüben algorithm on each processor.
- Various treatments possible at processor boundaries.
- Yields processor dependent coarsenings, and will not produce the same results for different numbers of processors.

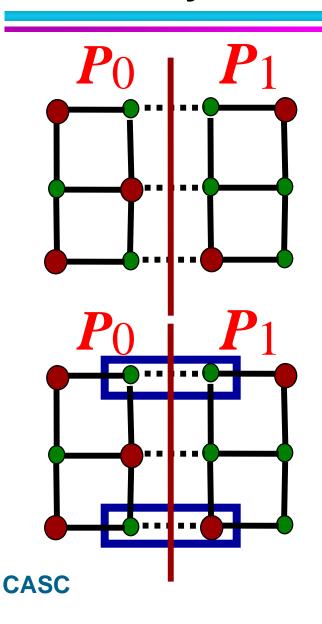


Measures

- Measure = number of strong influences.
- Possible treatments of the measure for parallel Ruge-Stüben coarsening:
 - Determine measures locally, no communication between processors (RS)
 - Use the 'correct' measures, i.e., take into account off-processor connections (RScm)

Parallel RS coarsening: boundary treatment: RS





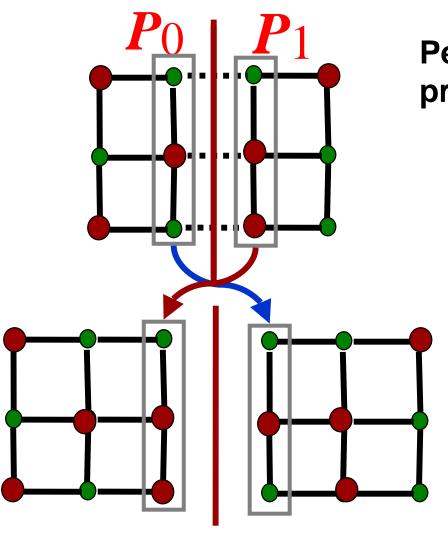
Perform first and second passes on each processor

Method 1: Do nothing.
Accept the coarsening provided by the independent processors.

Problem: Leaves $F \Leftrightarrow F$ dependencies without mutual C-points

Parallel RStüben coarsening: boundary treatment (RS2b)





Perform first pass on each processor

Perform second pass locally on each processor, augmented by boundary points from neighbor

Choices must be made about how to resolve conflicting decisions among processors

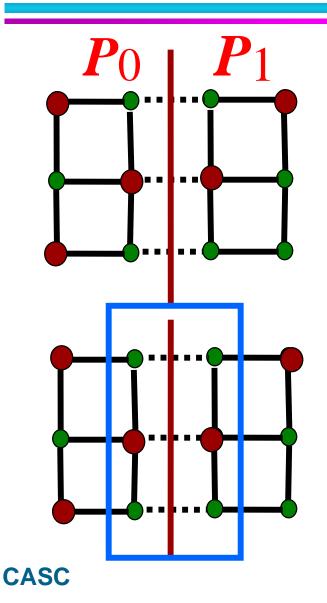


Boundary conflict resolution

- Methods to resolve conflicting coarsenings at processor boundaries:
 - Largest processor ID wins (RS3, RS2b) may violate (C1)
 - keep all coarse points (RS3c)
 does not violate (C1)
 may yield "too many" coarse-points
 giving high operator complexity

Parallel RS coarsening: boundary treatment (RS3)





Perform first and second pass on each processor

Perform a third pass, (a second "second pass"), only on those points adjacent to processor boundaries

Choices must be made about how to resolve conflicting decisions among processors

Parallel Ruge-Stüben coarsening results



7 pt 3D Laplacian		
Procs.	Setup	Op. Cplx
1	20	4.91
2	30	5.25
4	48	5.71
8	79	6.23
16	119	6.75
32	194	6.98
64	360	7.34
128		
Procs	Solve	C.F.
1	36	0.065
2	40	0.081
4	43	0.111
8	48	0.210
16	389	0.246
32	3433	0.605
64	3352	0.384

Ruge-Stüben coarsening is much faster and yields much better complexities than Cleary-LJP on the 7-pt Laplacian

Note that the solve times jump by orders of magnitude as problem grows. Parallel Ruge leads to large "coarsest" grids with direct solve.

Solution: hybrid coarsening?

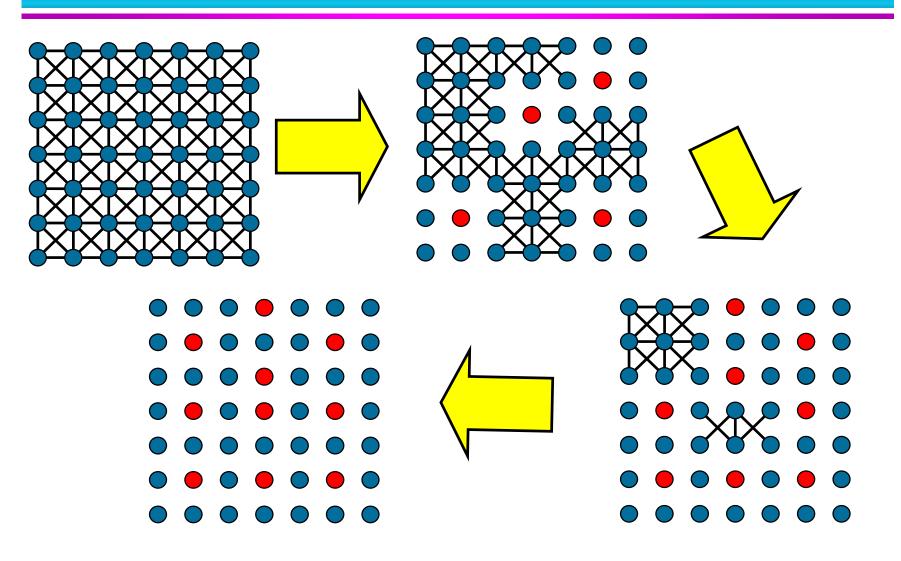
A new approach: the Cleary-LJP algorithm



- The Ruge-Stüben algorithm is inherently sequential.
- A new algorithm was proposed by Andrew Cleary, following parallel-independent-set algorithms developed by Luby and later by Jones & Plasssman
- Resulting coarsening algorithm (Cleary-LJP) is fully parallel, independent of the number of processors or processor topology. Serial prototype early 98, parallel code late 98.

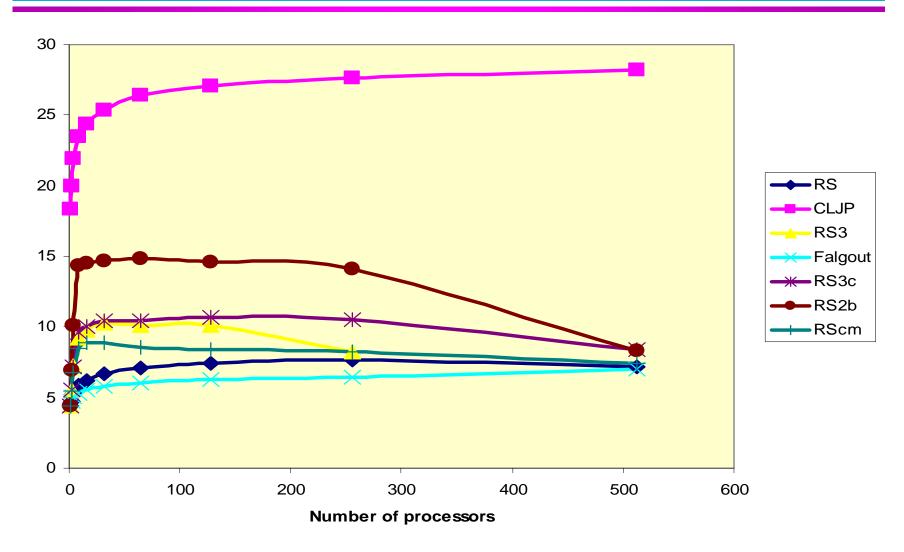
The C-LJP coarsening is fully parallel; independent of *P*





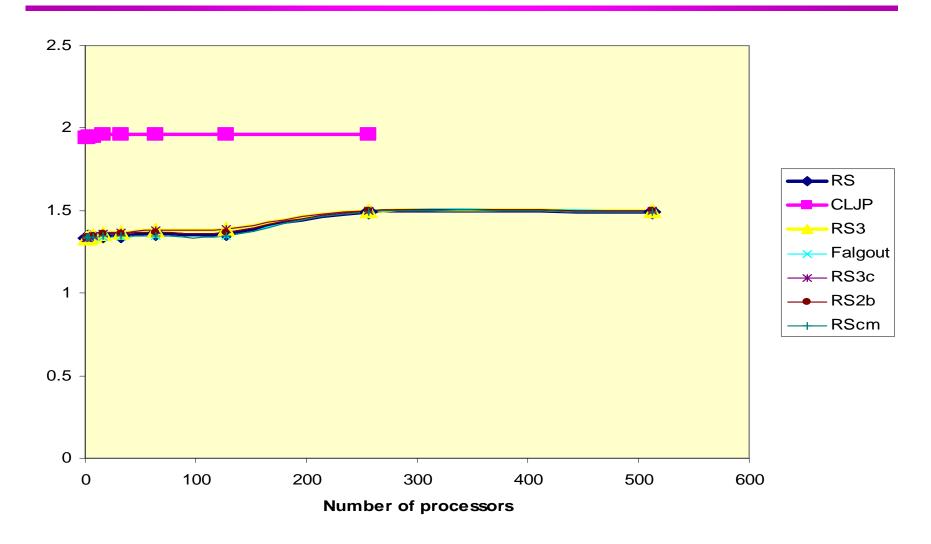
Operator Complexities





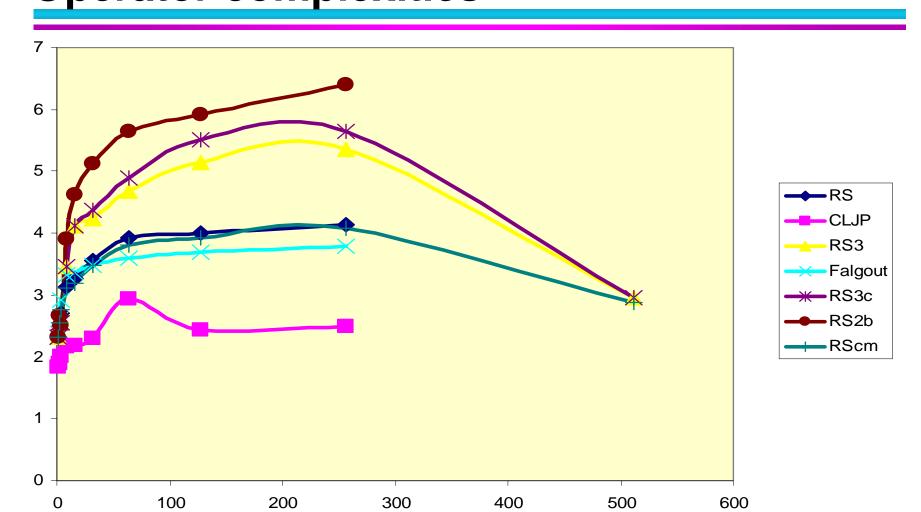
9 point 2D Laplacian: Operator complexities





27 point 3D Laplacian: Operator complexities

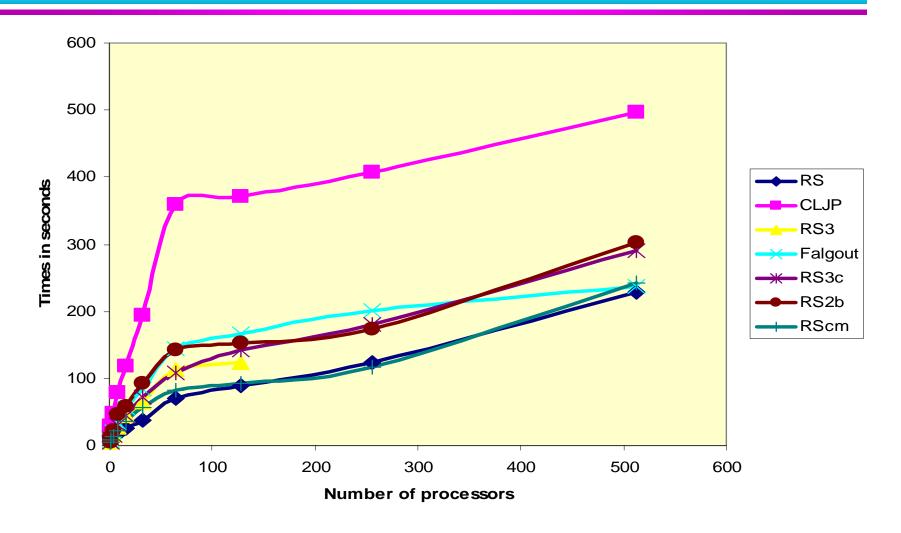




Number of processors

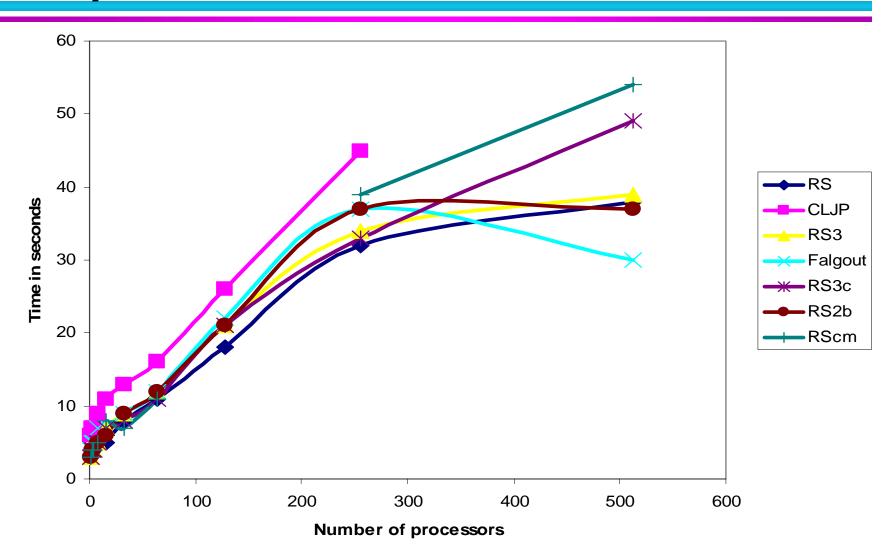
Setup times





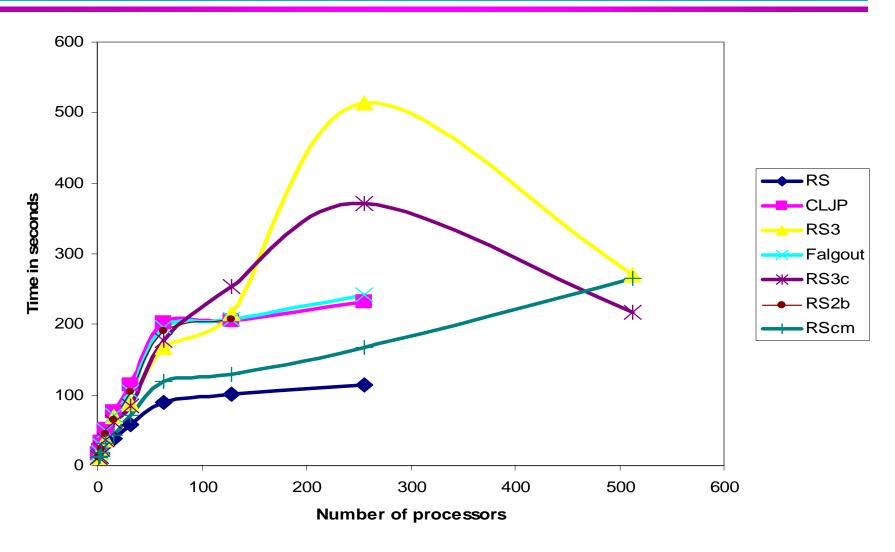
Setup times





Setup times







Relaxation techniques

Jacobi or weighted Jacobi

$$x^{(n+1)} = (2-w)x^{(n)} + wD^{-1}(b - Ax^{(n)})$$

e.g.
$$w = \frac{1}{\|D^{-1/2}AD^{-1/2}\|}$$

Gauß-Seidel

$$(D-L)x^{(n+1)} = Ux^{(n)} + b$$

where

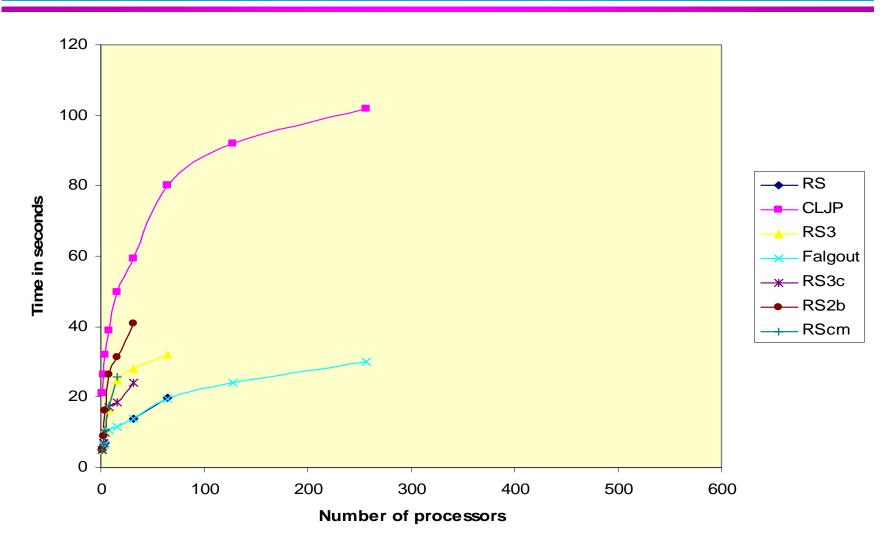
$$A = D - L - U$$

chaotic Gauß-Seidel

use new values when available, old values on processor boundaries

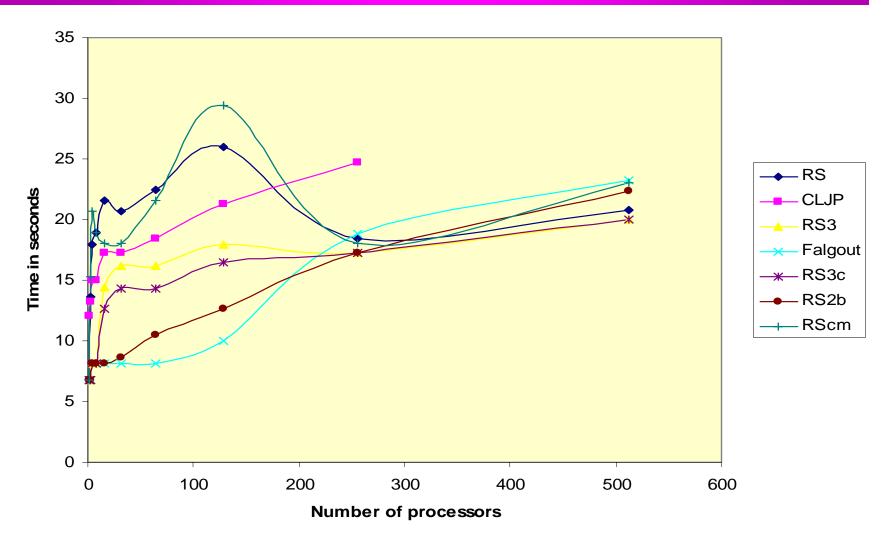
7 point 3D Laplacian: Solve times chaotic GS





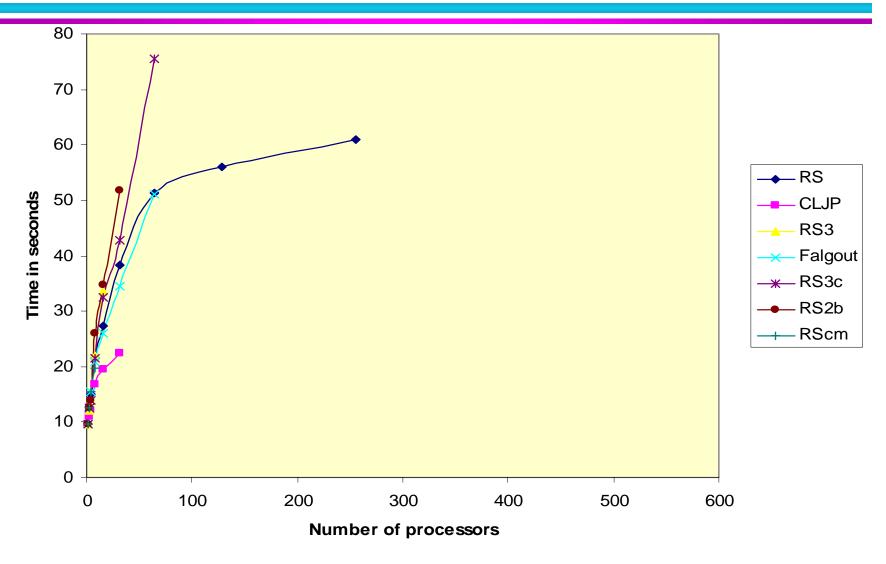
Solve times chaotic GS





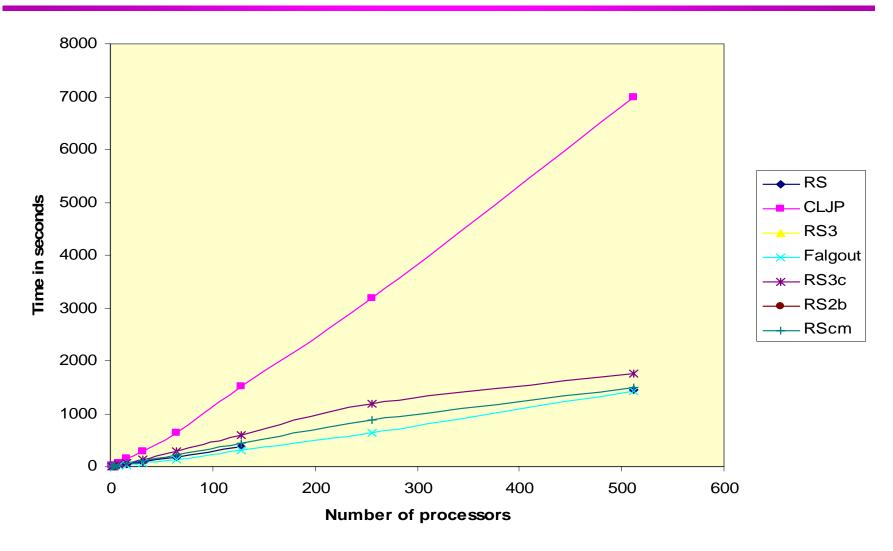
27 point 3D Laplacian: Solve times chaotic GS





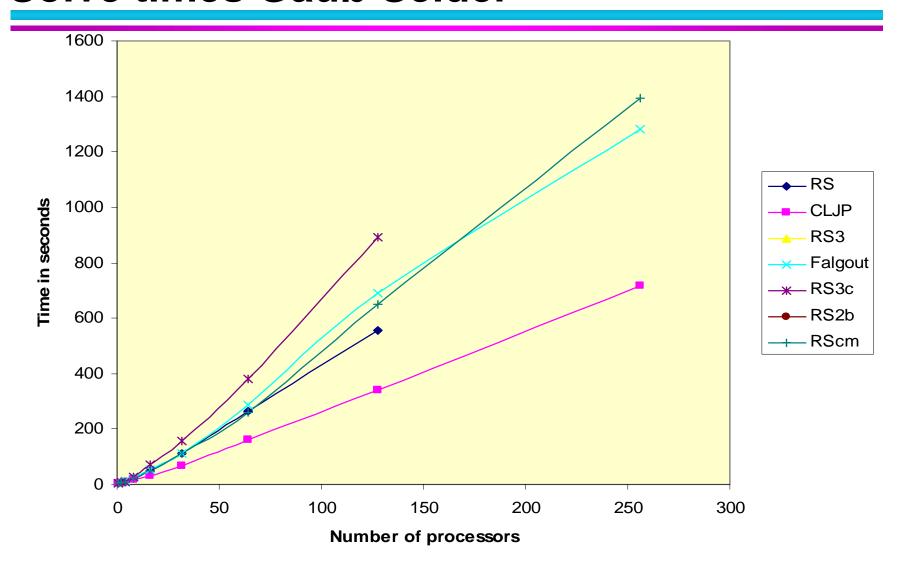
7 point 3D Laplacian: Solve times Gauß-Seidel





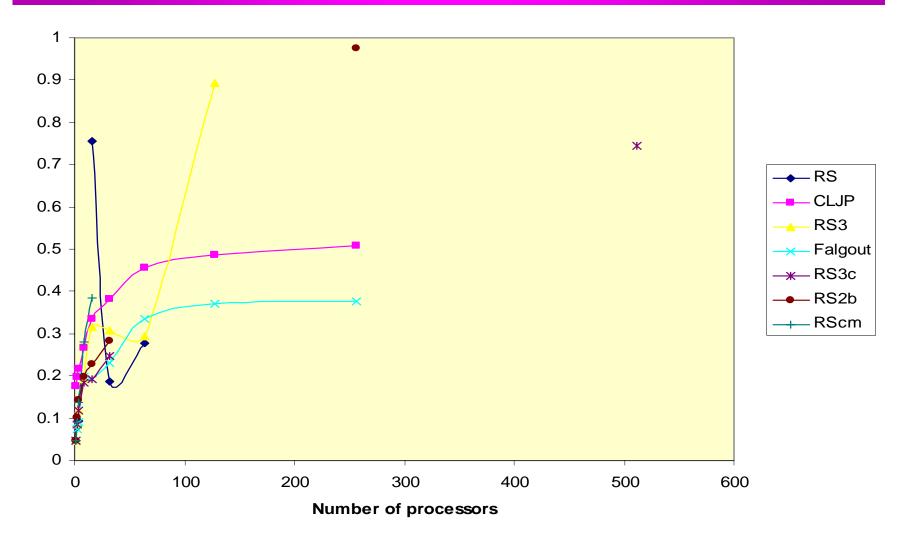
27 point 3D Laplacian: Solve times Gauß-Seidel





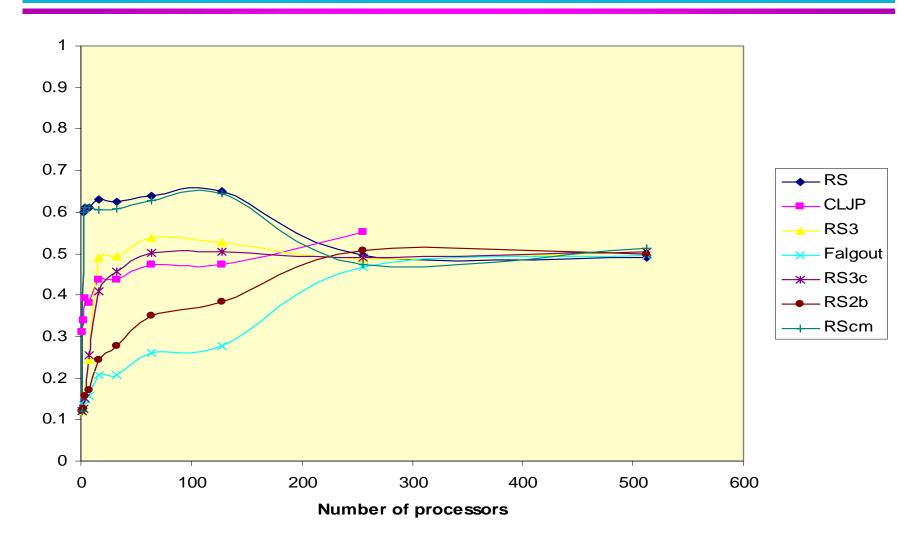
As. Conv. Factor chaotic GS





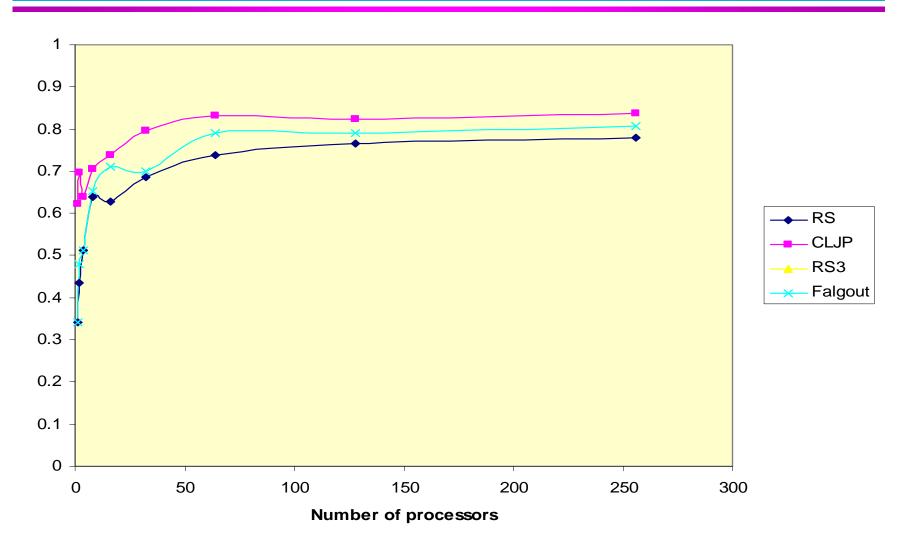
As. Conv. Factor chaotic GS





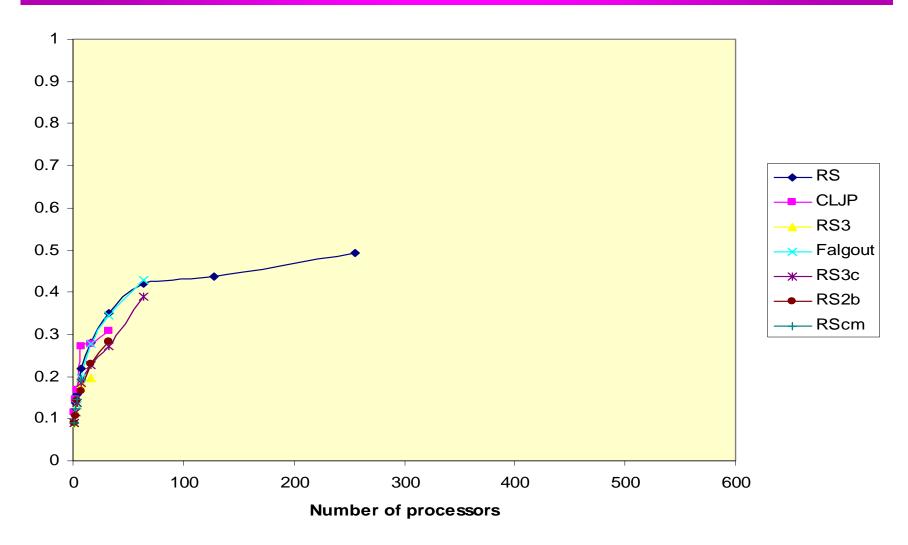
As. Conv. Factor wt. Jacobi





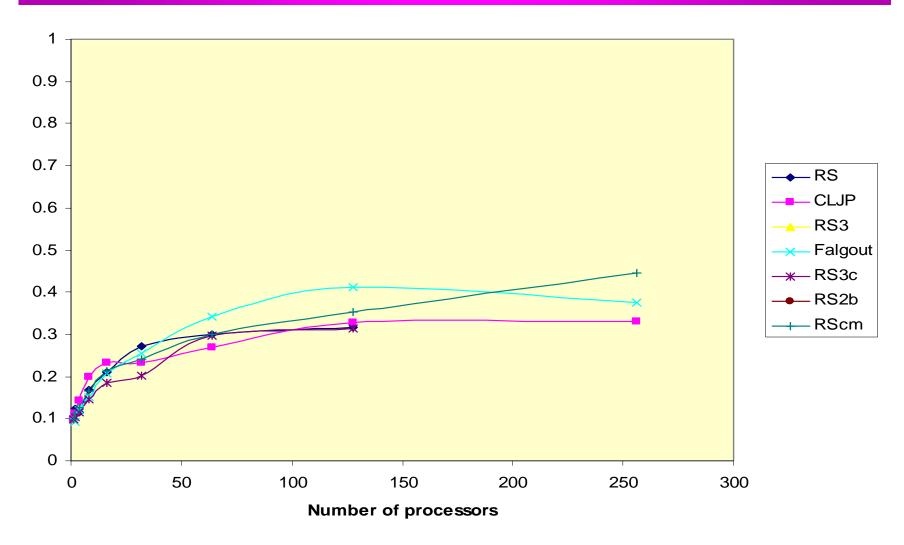
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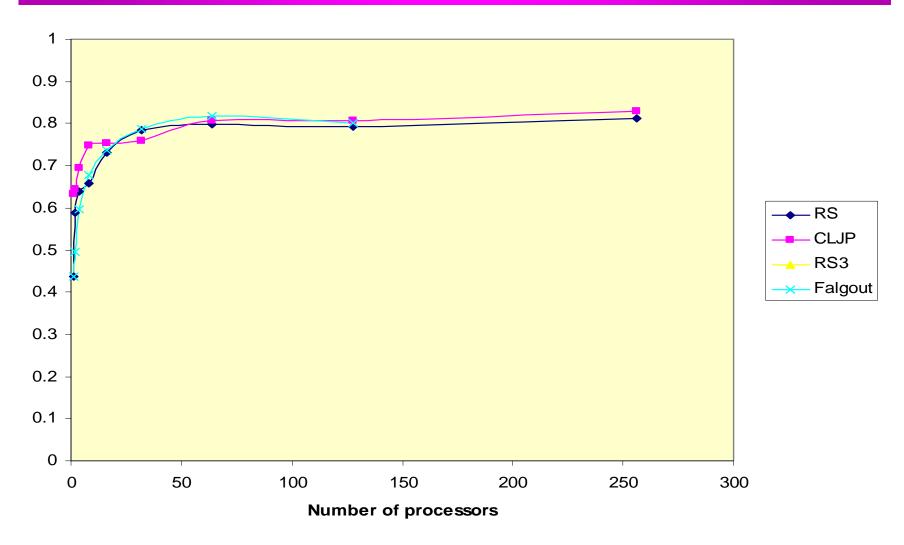
As. Conv. Factor GS





As. Conv. Factor wt. Jacobi







Conclusions & Future Work

- AMG has been parallelized. It shows reasonably good scalability.
- Testing is still needed to implement the algorithms efficiently; to determine better ways of treating processor boundaries, operator complexities, and growing convergence factors.
- Future computer science plans include load balancing and efficient cache useage.
- Future algorithmic development centers on implementing "system" solvers and determining MG components using the finite-element stiffness matrices
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